

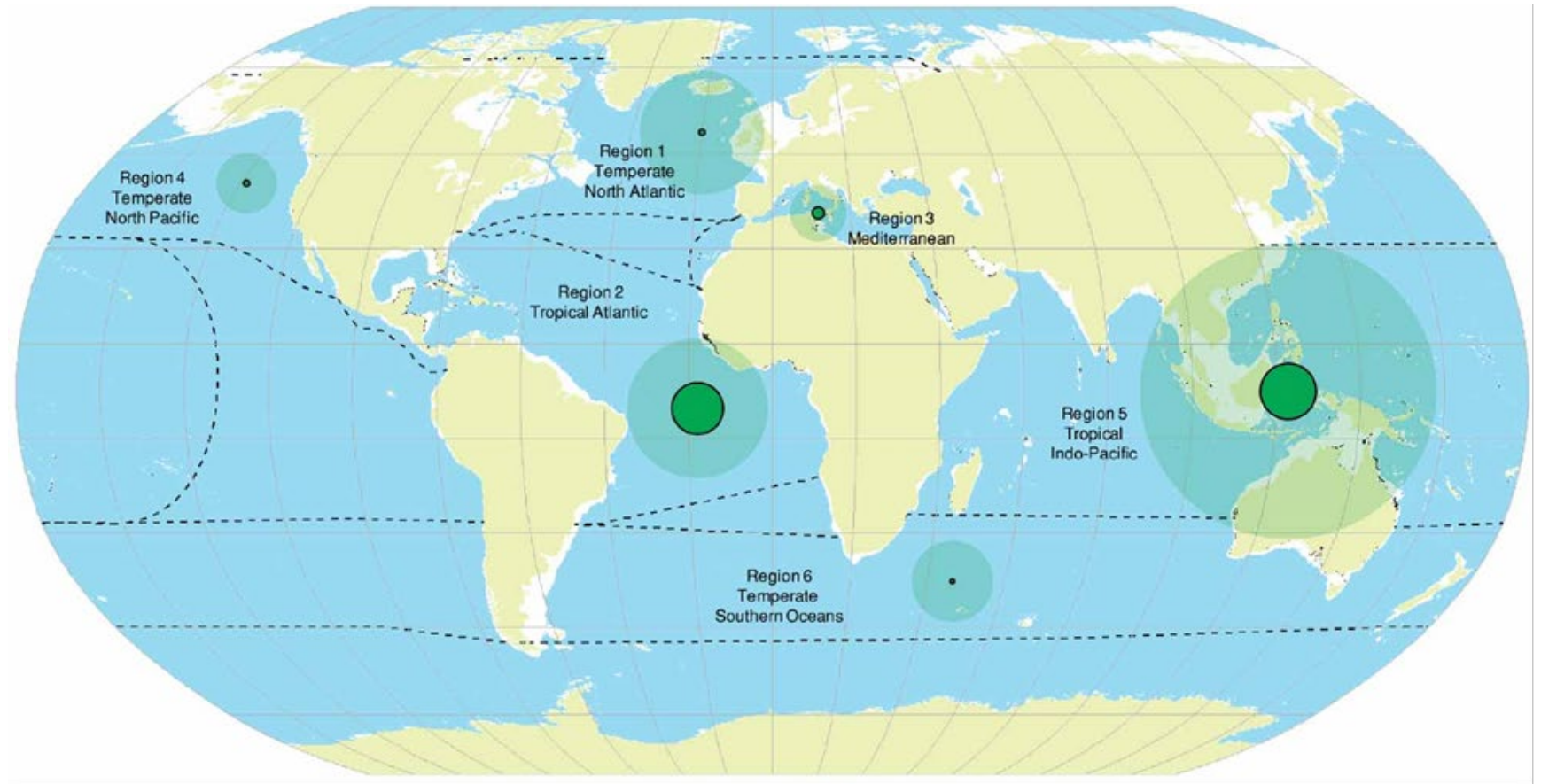
Providing a Framework for Seagrass Mapping in Coastal Systems Using High Spatial Resolution Satellite Imagery



Megan Coffey, Ph.D.
NOAA GST
megan.coffey@noaa.gov

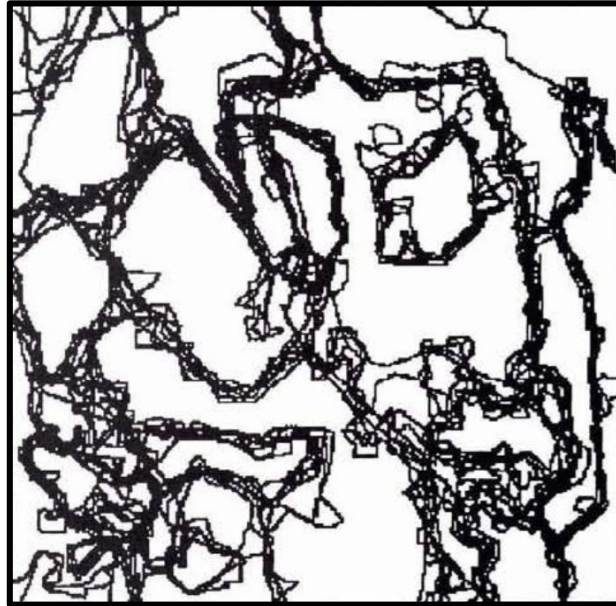






- MaxEnt predicted seagrass relative area
- Actual mapped seagrass relative area

Estimated global seagrass coverage ranges between 150,000 and 4,320,000 km²

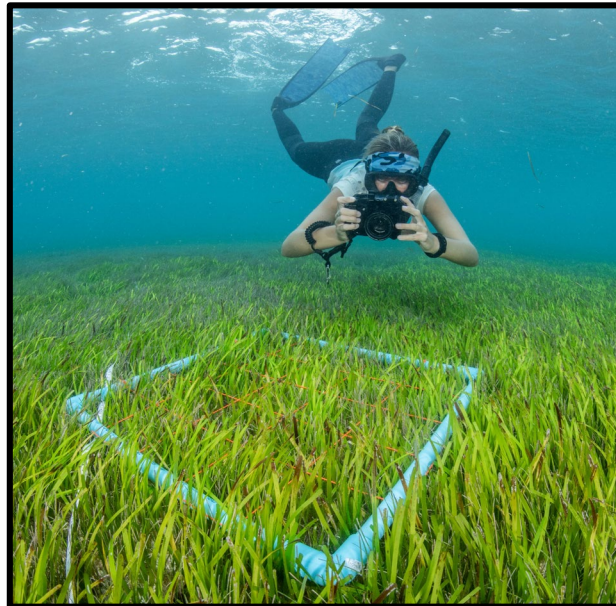


Aerial photointerpretation

Inconsistent results

Cost prohibitive

Time consuming

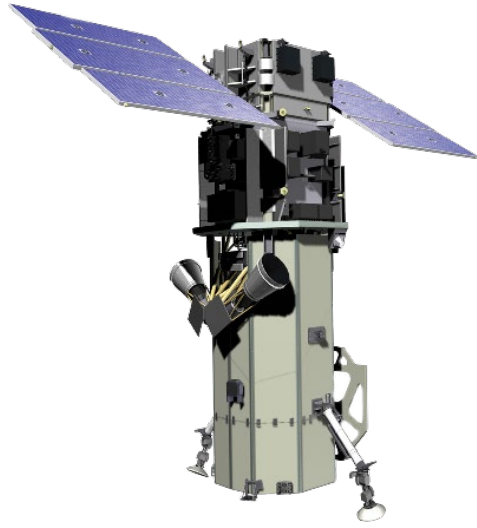


Field quadrats

Low spatial coverage

Time consuming

Many areas inaccessible



Maxar's WorldView-2

Lifespan: 2009 to present

Spatial resolution: 1.84 m

Spectral resolution: 6 visible, 2 NIR

Temporal resolution: inconsistent & unpredictable



Maxar's WorldView-3

Lifespan: 2014 to present

Spatial resolution: 1.24 m

Spectral resolution: 6 visible, 2 NIR

Temporal resolution: inconsistent & unpredictable



Maxar's WorldView-2

Lifespan: 2009 to present

Spatial resolution: 1.84 m

Spectral resolution: 6 visible, 2 NIR

Temporal resolution: inconsistent & unpredictable



Maxar's WorldView-3

Lifespan: 2014 to present

Spatial resolution: 1.24 m

Spectral resolution: 6 visible, 2 NIR

Temporal resolution: inconsistent & unpredictable

Most studies focus on a single study area, limiting reproducibility

Process imagery

Radiometric
calibration

```
graph TD; A[Radiometric calibration] --> B[Atmospheric correction]; B --> C[Orthorectification and mosaicking];
```

Atmospheric
correction

Orthorectification
and mosaicking

Process imagery

Radiometric calibration



Atmospheric correction



Orthorectification and mosaicking

Classify seagrass

Define regions of interest



Apply deep convolutional neural network

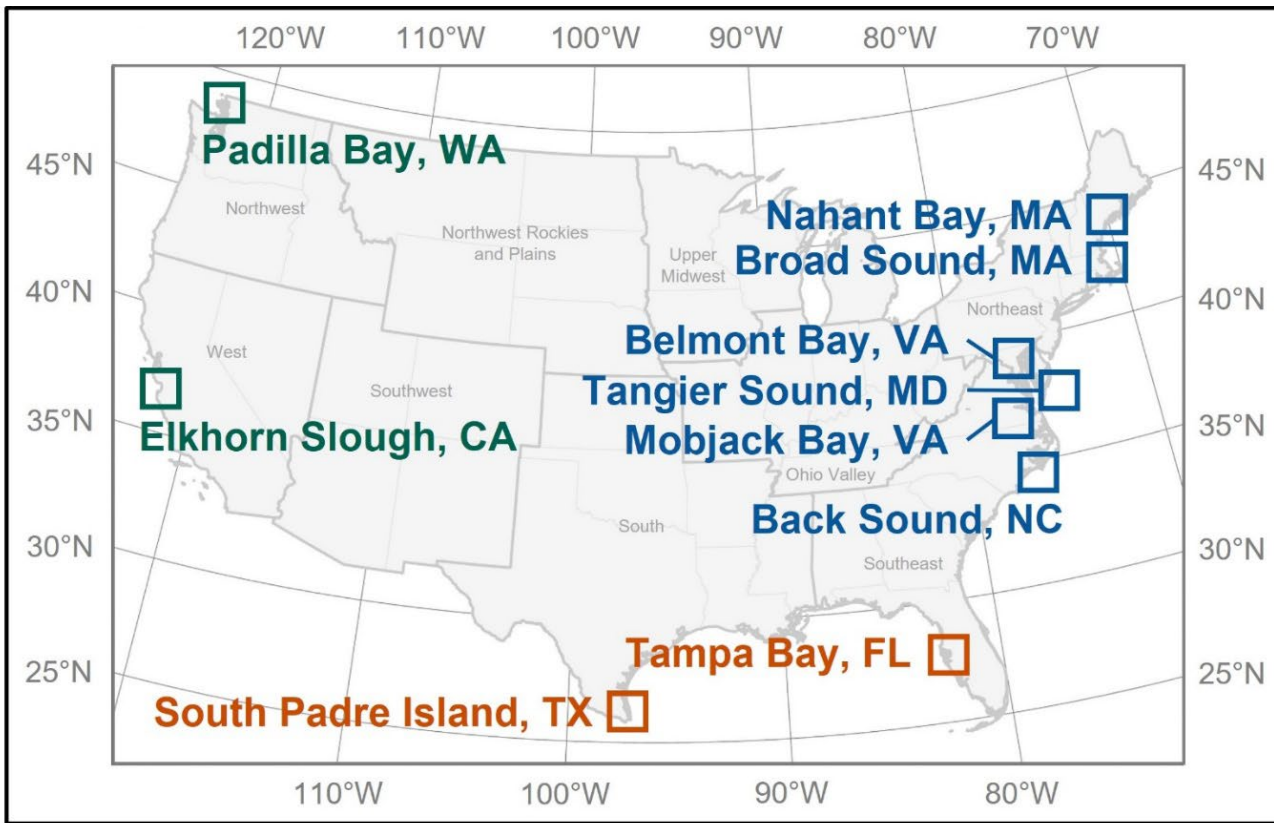


Compare to photointerpreted aerial imagery

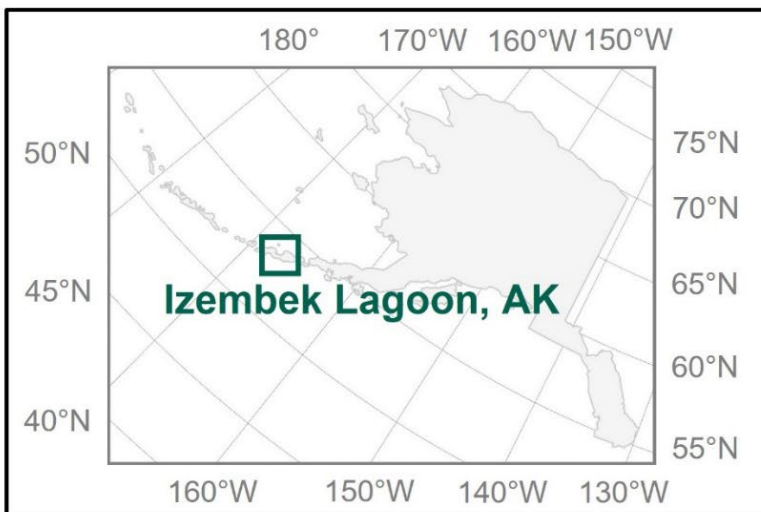
First reproducible processing regime for WorldView-2 & WorldView-3

Tested at St. Joseph Bay, FL, with 97% overall agreement

*How well does our image processing
and seagrass classification regime
first defined in Coffey et al. (2020)
perform in more complex optical
environments?*



***11 study sites, representing
3 of 6 global seagrass
bioregions and each coastal
United States climate region***



Seagrass bioregions:
Temperate North Pacific
Tropical Atlantic
Temperate North Atlantic

For each study area

Reference data

*Obtain seagrass
reference data from
local mapping efforts*

Satellite imagery

*Acquire spatially
coincident
Worldview-2 or
WorldView-3 imagery,
targeting minimal
temporal offset and
visually ideal water
clarity*

Satellite processing

*Following
Coffer et al. (2020),
apply preprocessing
and image
classification*

Assess agreement

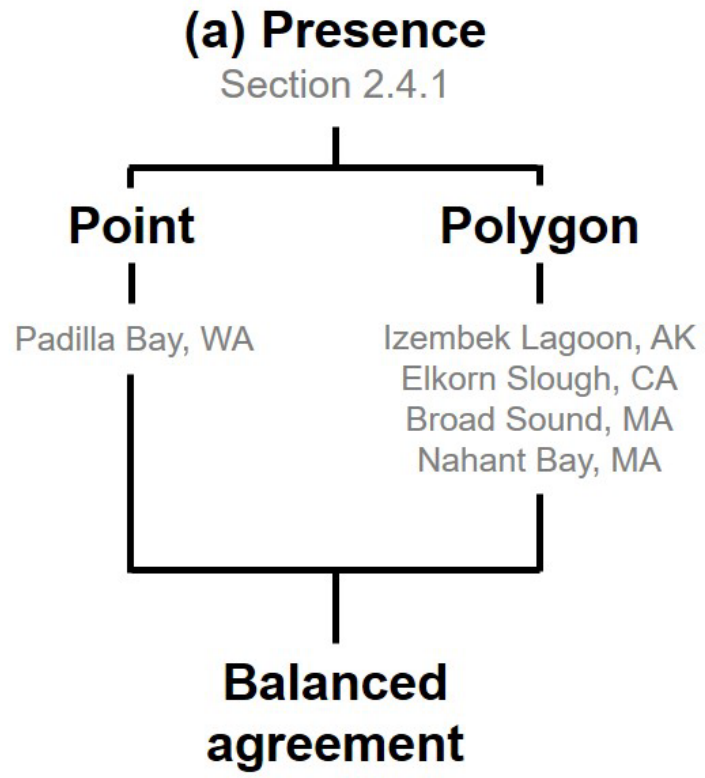
*Depending on
reference data format,
between reference
data and satellite
image classifications*

*Reference data
classification type*

*Reference data
spatial data type*

Study areas

Statistical test

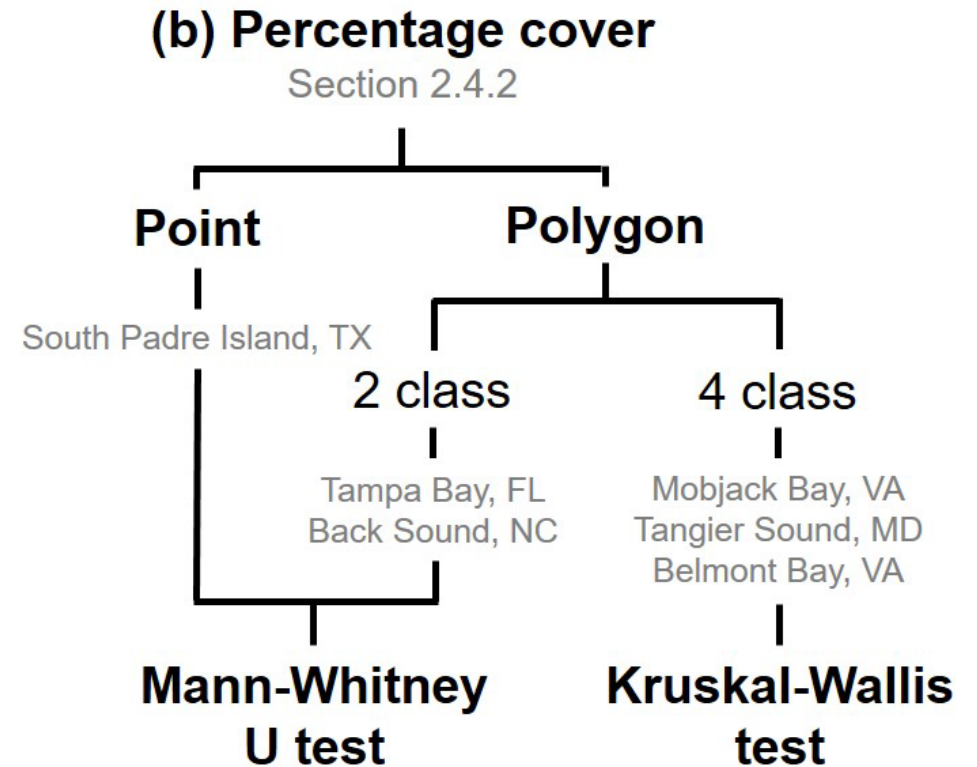
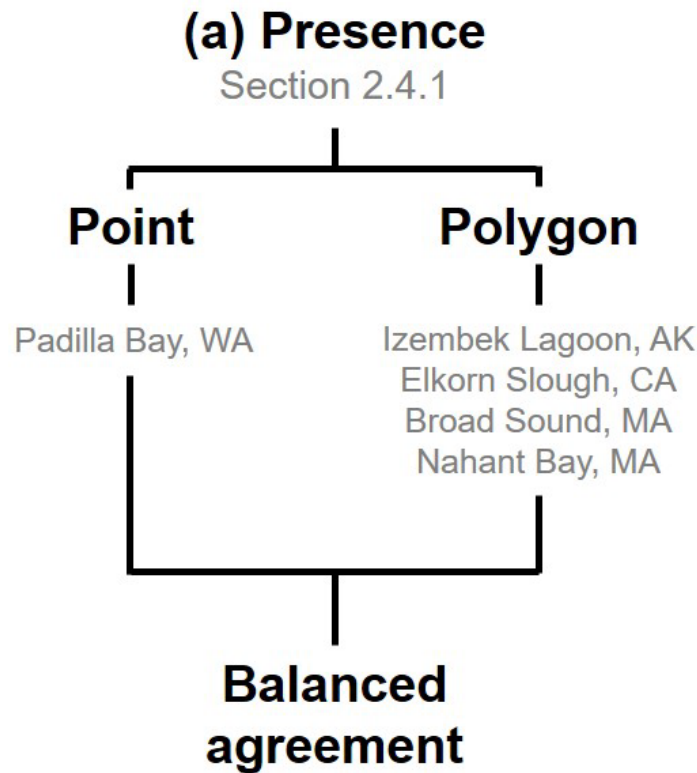


Reference data classification type

Reference data spatial data type

Study areas

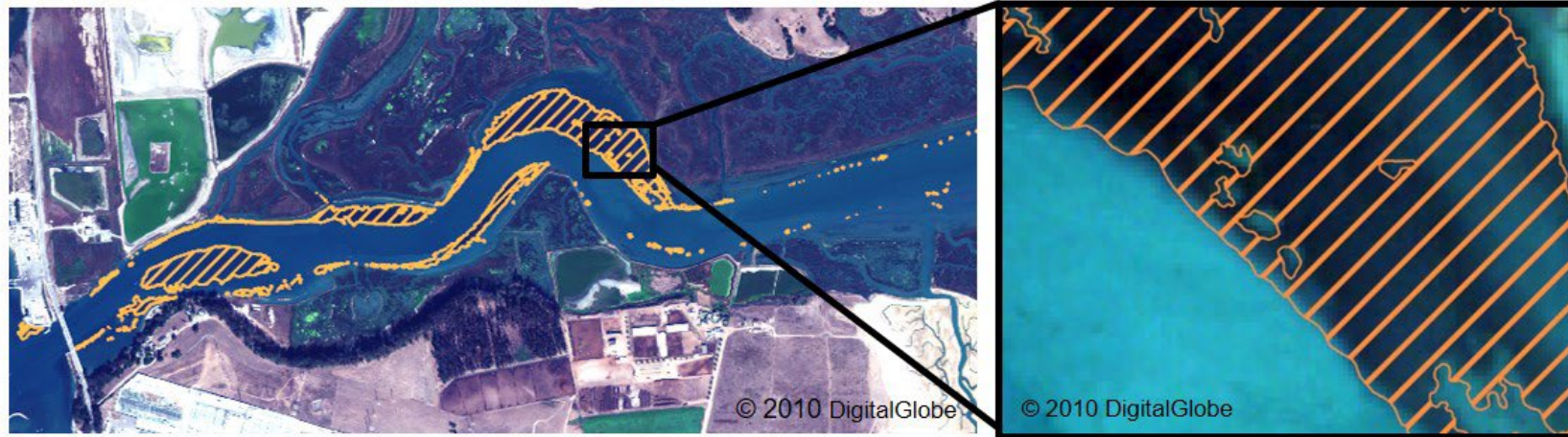
Statistical test



Novel statistical approaches for comparing pixel-based satellite data to polygon-based reference data

Elkhorn Slough, CA

(a) WorldView-2 image with field data



(b) WorldView-2 image classification



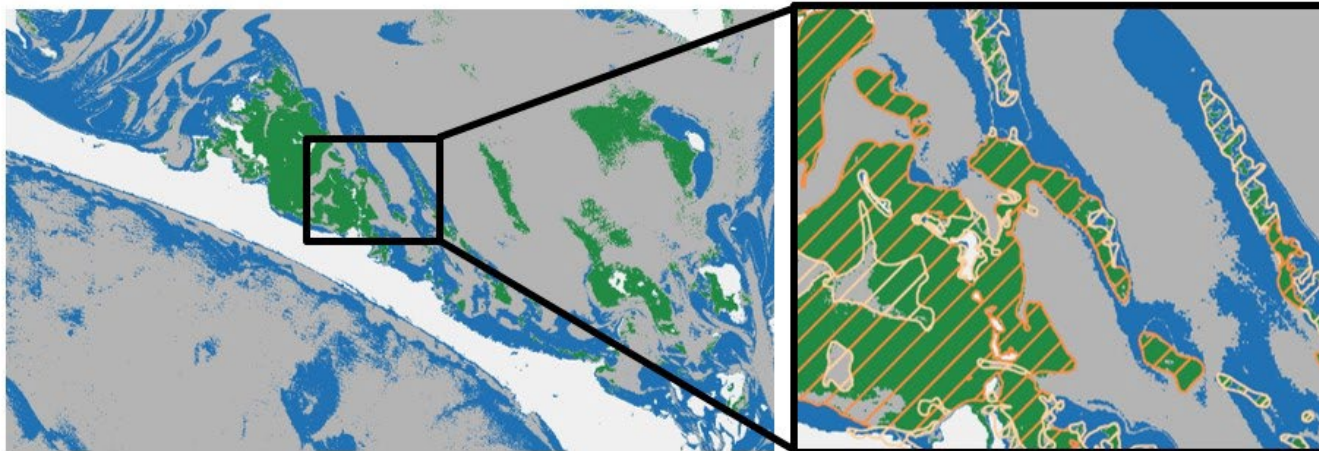
86% balanced agreement (73% sensitivity; 100% specificity)

Back Sound, NC

(a) WorldView-2 image with field data



(b) WorldView-2 image classification



Satellite image classification

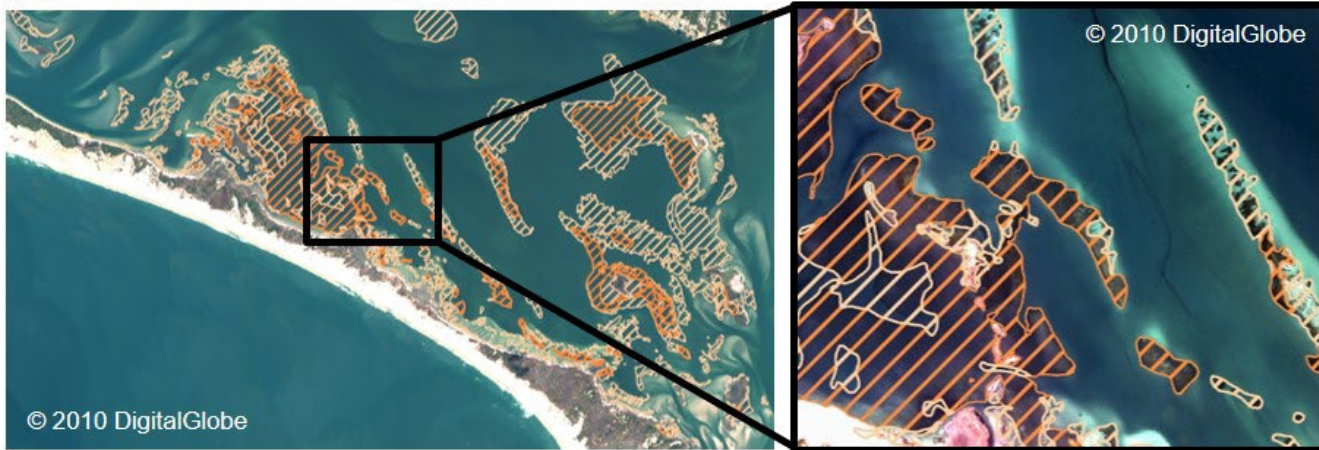
- Land
- No data
- Seagrass
- No seagrass

Reference data

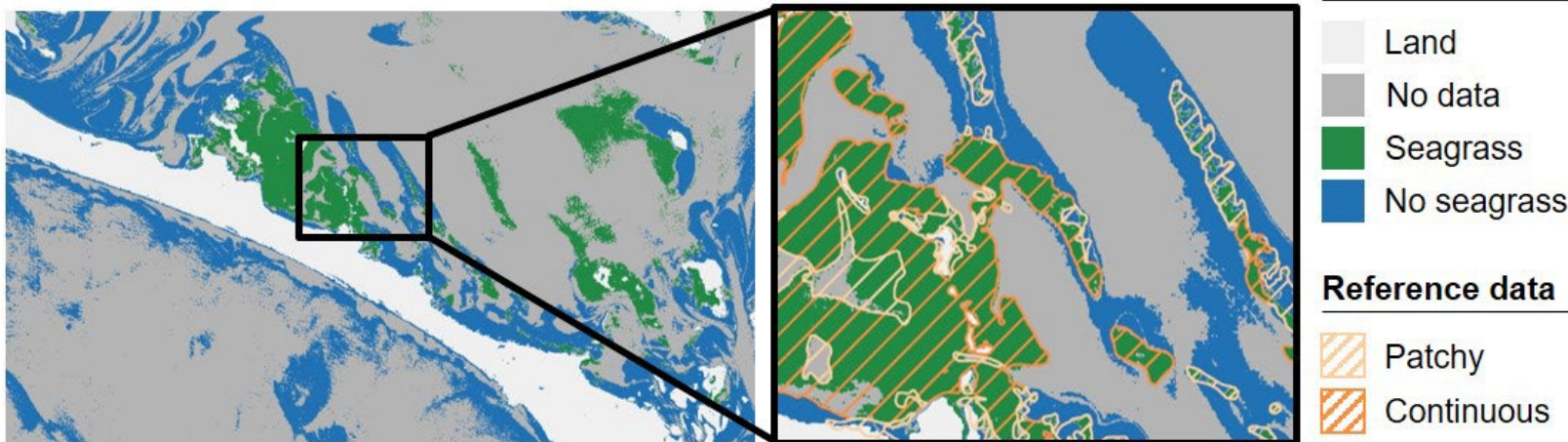
- Patchy
- Continuous

Back Sound, NC

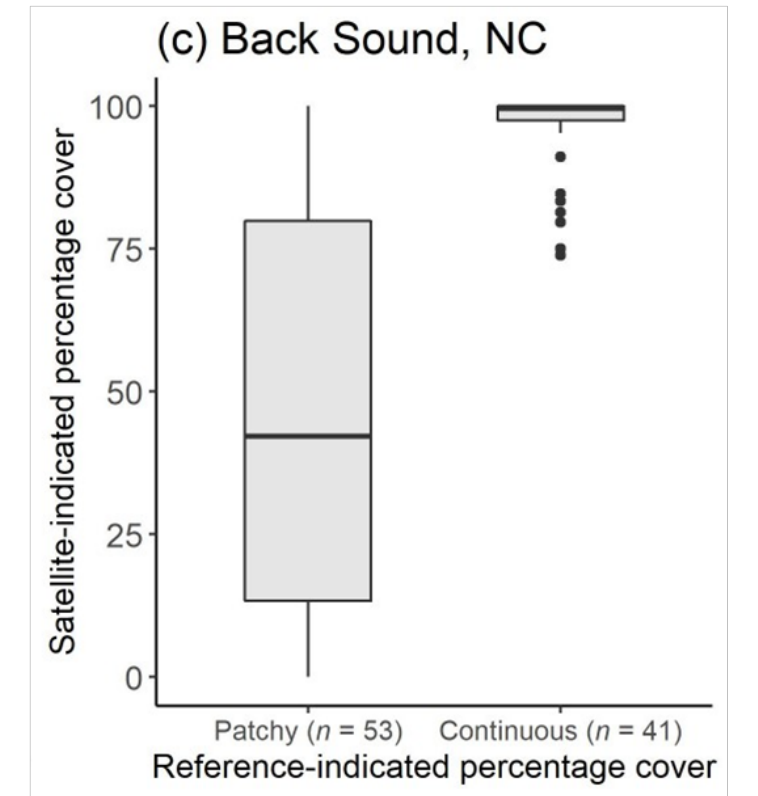
(a) WorldView-2 image with field data



(b) WorldView-2 image classification



M = 42% M = 100%
 $r_{tb} = 0.83$; large association between datasets



Large associations indicate agreement between datasets



Satellite data misclassifications

Sparse seagrass: Likely underestimated as bright sand dominates

Spectral similarity: Algae and seagrass are often intermixed, and all substrates had similar spectral features at depth

Low SNR: WorldView-3 may be less appropriate for aquatic applications

(Coffer et al., 2022 *Int. J. Remote Sens.*)



Satellite data misclassifications

Sparse seagrass: Likely underestimated as bright sand dominates

Spectral similarity: Algae and seagrass are often intermixed, and all substrates had similar spectral features at depth

Low SNR: WorldView-3 may be less appropriate for aquatic applications

(Coffer et al., 2022 *Int. J. Remote Sens.*)



Reference data misclassifications

Sparse seagrass: Overestimated as sand and seagrass are amalgamated (Meehan et al., 2005 *Estuaries*)

Historic and potential habitat: Photointerpretation often includes buffers around existing seagrass beds to account for historic conditions or potential habitat



Satellite data misclassifications

Sparse seagrass: Likely underestimated as bright sand dominates

Spectral similarity: Algae and seagrass are often intermixed, and all substrates had similar spectral features at depth

Low SNR: WorldView-3 may be less appropriate for aquatic applications
(Coffer et al., 2022 *Int. J. Remote Sens.*)



Reference data misclassifications

Sparse seagrass: Overestimated as sand and seagrass are amalgamated (Meehan et al., 2005 *Estuaries*)

Historic and potential habitat: Photointerpretation often includes buffers around existing seagrass beds to account for historic conditions or potential habitat



Dataset mismatches

Temporal offset: Ranged from 1 month to 16 years

Spatial offset: Raster data being compared to point and polygon data



Agreement between satellite and reference data

Balanced agreement ranged from 58% to 86%, with better agreement for seagrass absence

Mann-Whitney U and Kruskal-Wallis tests demonstrated moderate to large correlations between datasets



Agreement between satellite and reference data

Balanced agreement ranged from 58% to 86%, with better agreement for seagrass absence

Mann-Whitney U and Kruskal-Wallis tests demonstrated moderate to large correlations between datasets

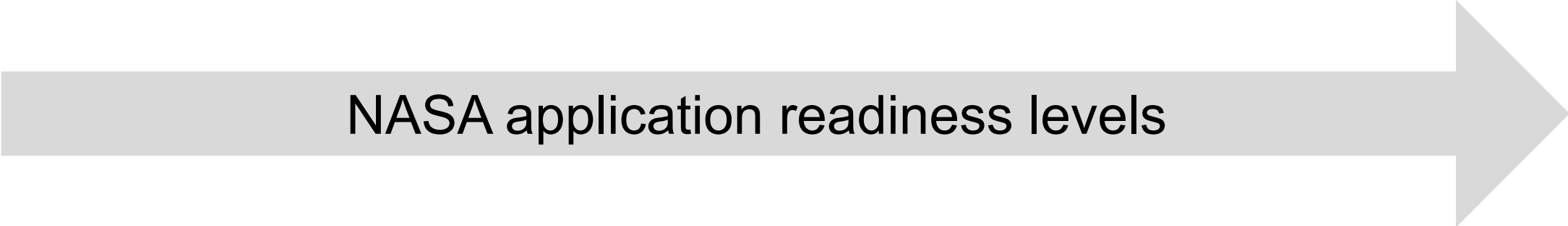


Regional and global mapping

The same methods can be applied across varying seagrass bioregions, atmospheric conditions, and optical water types

Marks a significant step toward developing a consistent, operational approach for mapping seagrass coverage at large scales

APPLICATIONS



NASA application readiness levels

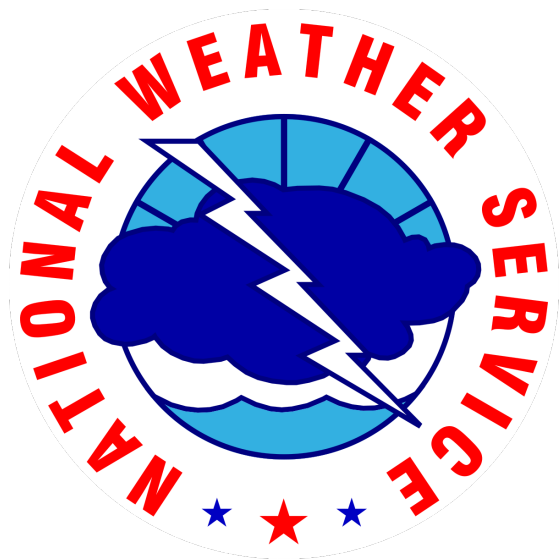
Phase 1:
Discovery and
feasibility

Phase 2:
Development,
testing, and
validation

Phase 3:
Integration
into partner's
system



1. **Numerical weather prediction models:** *Initialized using satellite imagery and local observations*
2. **Local forecasts generated:** *Model output analyzed and scrutinized using localized, individualized scientific expertise.*



1. **Numerical weather prediction models:** *Initialized using satellite imagery and local observations*
2. **Local forecasts generated:** *Model output analyzed and scrutinized using localized, individualized scientific expertise.*

Methods presented here should be used in the same manner. Satellite data and local knowledge are used to estimate regional seagrass coverage which is then tailored by local experts.